

Topics in Response Surface Model Adequacy Assurance and Assessment

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An Alternative Concept of *Quality* in Experimental Aeronautics

- Traditional concept of quality in wind tunnel testing
 - Data-centric: "Quality" means "Data Quality" in traditional testing
 - Associated with low levels of unexplained variance in a data sample
- An alternative concept of quality
 - Introduced to the Langley experimental aeronautics community in the mid-90's as the Modern Design of Experiments (MDOE)
 - Associated with inference error probability
 - "Quality" means "getting the right answer"
 - Low probability of inference error
 - Independent of quality of the data



Response Surface Modeling

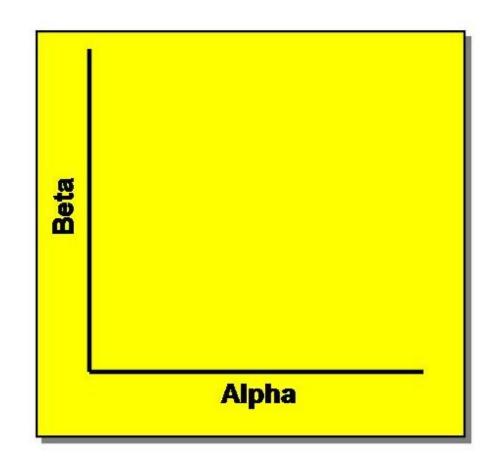
- Response Surface Models are mathematical functions representing responses (forces/moments, etc.) as a function of independent variables (AoA, Mach No., etc.)
- Quality is cast in terms of modeling adequacy
 - For an adequate model, no more than a specified percentage of response predictions are outside acceptable tolerance limits
 - Quality, or model adequacy, must be both assured and assessed
- Model adequacy is assured through the design by
 - Data volume specification (How many points)
 - Site selection within the design space (Which points)
 - Number and selection of points to be replicated
 - Order in which the points are acquired

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Model adequacy is assessed by examination of residuals

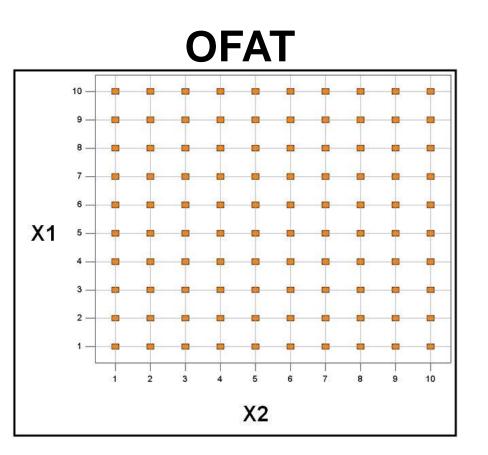
An Inference Space

- A Coordinate System
- One axis for each variable
- Each point represents a unique combination of variable levels
- A response surface is constructed "over" an inference space

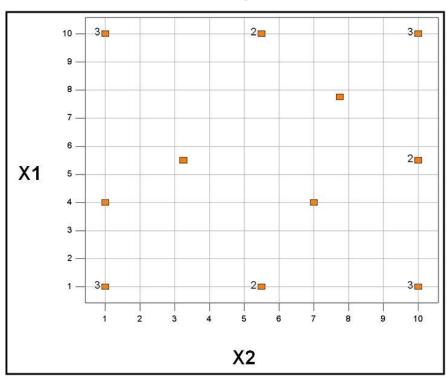




Design Space Comparisons

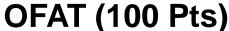


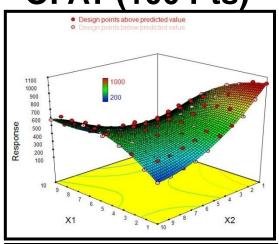
MDOE

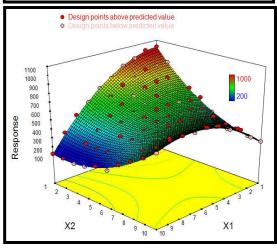




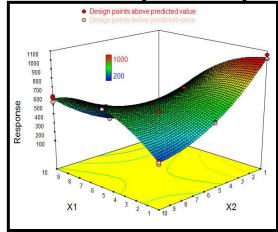
OFAT and MDOE Response Surfaces

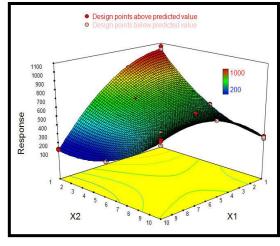






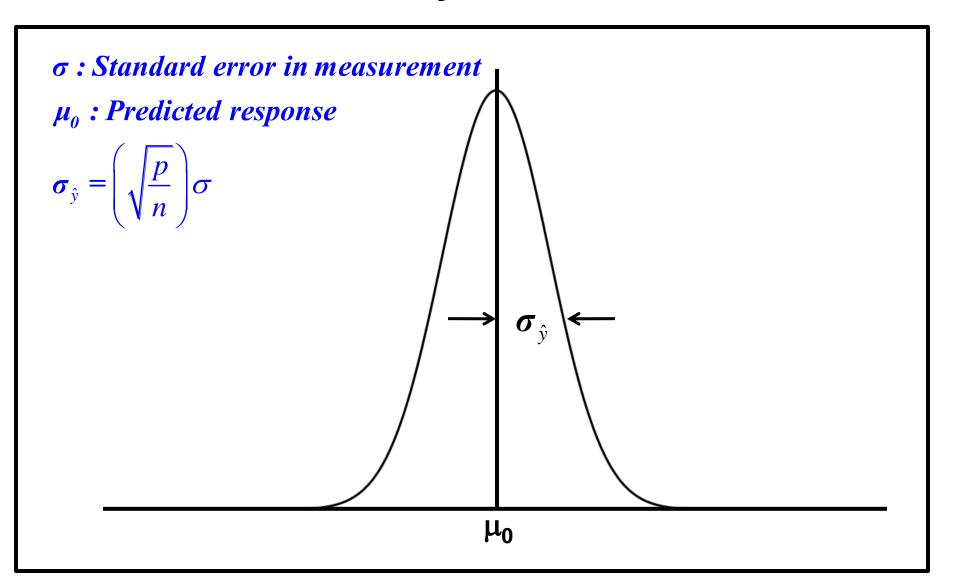
MDOE (22 Pts)





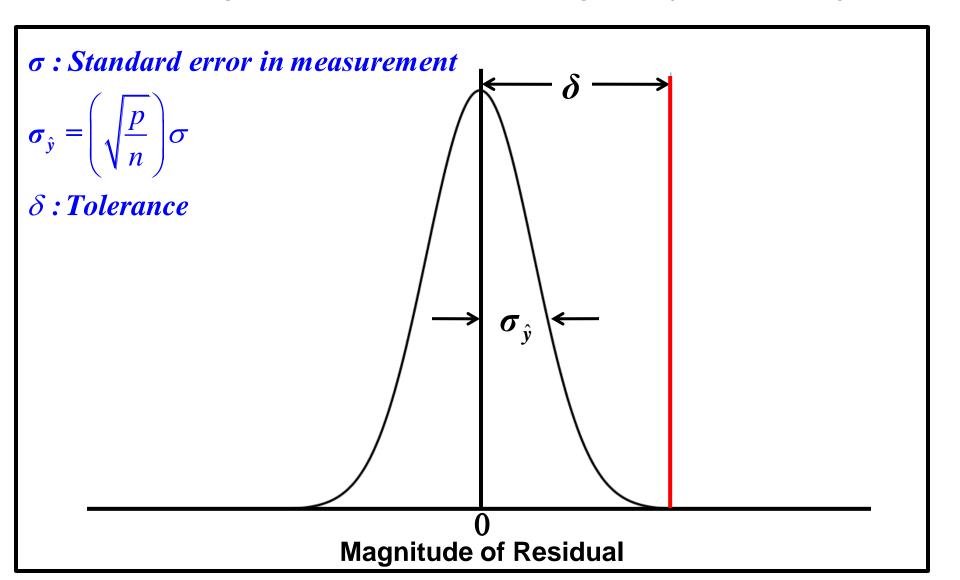


The Mathematics of Quality Assurance and Quality Assessment



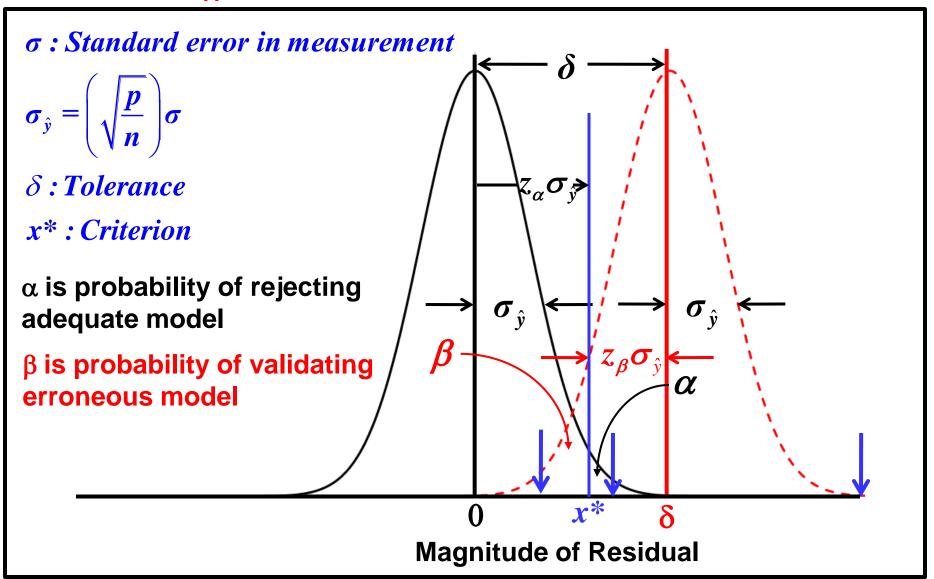
Reference Distribution Under H₀

*H*₀: Null hypothesis that there is no difference between predicted and measured response (Residual is 0)



Reference Distributions for Residuals Black – H₀: True residual is zero

Red – H_A: True residual is borderline unacceptable



Data Volume Requirement

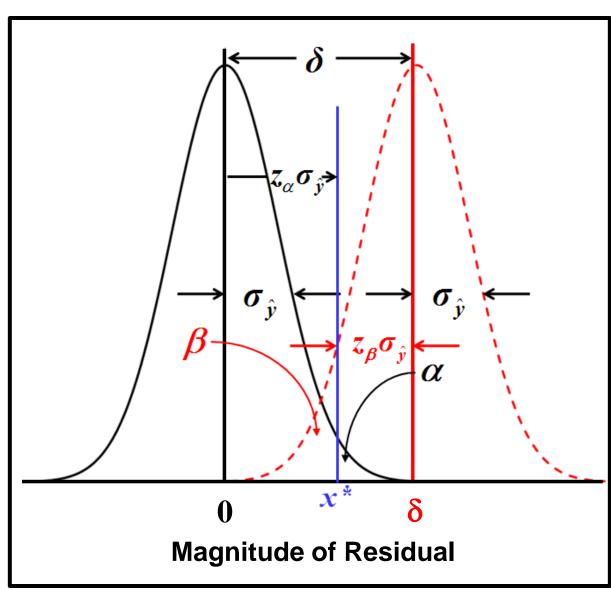
$$\delta = \left(z_{\alpha} + z_{\beta}\right)\sigma_{\hat{y}}$$

$$\sigma_{\hat{y}} = \left(\sqrt{\frac{p}{n}}\right)\sigma$$

$$\delta^2 = \frac{p(z_{\alpha} + z_{\beta})^2 \sigma^2}{n}$$

$$n = p \left[\left(z_{\alpha} + z_{\beta} \right)^2 \frac{\sigma^2}{\delta^2} \right]$$





Data Volume Formula Some Practical Difficulties

$$n = p \left[\left(z_{\alpha} + z_{\beta} \right)^{2} \frac{\sigma^{2}}{\delta^{2}} \right]$$

- The data volume formula depends on five quantities
 - Three (p, α , and β) can often be specified by the design consultant
 - The tolerance, δ , should be specified by the customer
 - The standard measurement error, σ , should be specified by the facility
- The customer often prefers to specify tolerance as a multiple of σ , rather than in absolute terms
 - A customer may feel comfortable saying his tolerance is "2σ"
 - He doesn't always feel he has to know what "σ" is to say this



Incorporating Tolerance in Data Volume Estimates

- Consider the general case, in which $\delta = K\sigma$, where K is a constant specified by the customer
- Note that a specific "K" may eventually evolve as an industry convention (about which more in a moment)

$$\delta = K\sigma$$

$$n = p\left(z_{\alpha} + z_{\beta}\right)^{2} \frac{\sigma^{2}}{\delta^{2}}$$

$$n = \left(\frac{z_{\alpha} + z_{\beta}}{K}\right)^{2} p$$



Special Case for Tolerance, δ

• We have for the general case in which $\delta = K\sigma$.

$$n = \left(\frac{z_{\alpha} + z_{\beta}}{K}\right)^{2} p$$

- Let δ = 95% LSD (Least Significant Difference)
 - This is the smallest difference between two replicated measurements that can be resolved with 95% confidence
 - It may be regarded as a reasonable tolerance specification

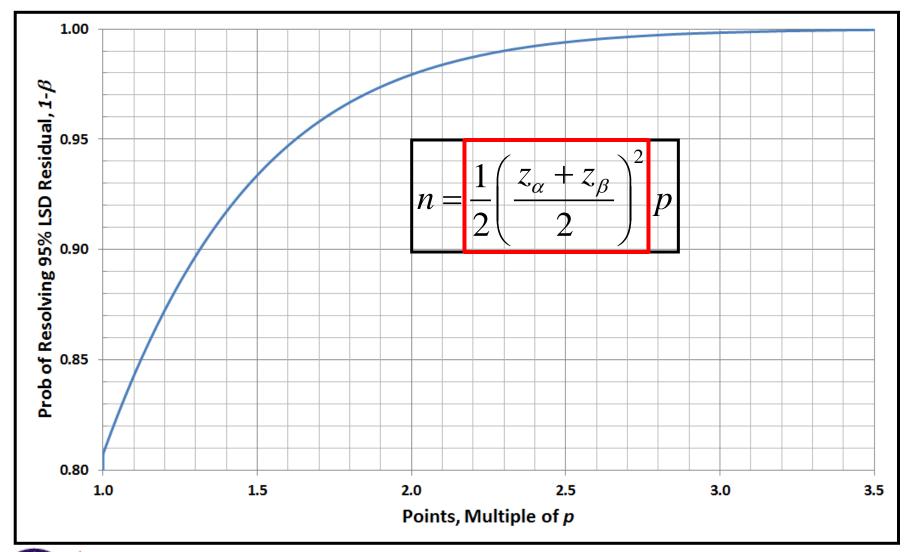
$$\delta = 95\% \text{ LSD} = 2\sqrt{2}\sigma \rightarrow K = 2\sqrt{2}$$

$$n = \frac{1}{2} \left(\frac{z_{\alpha} + z_{\beta}}{2} \right)^2 p$$



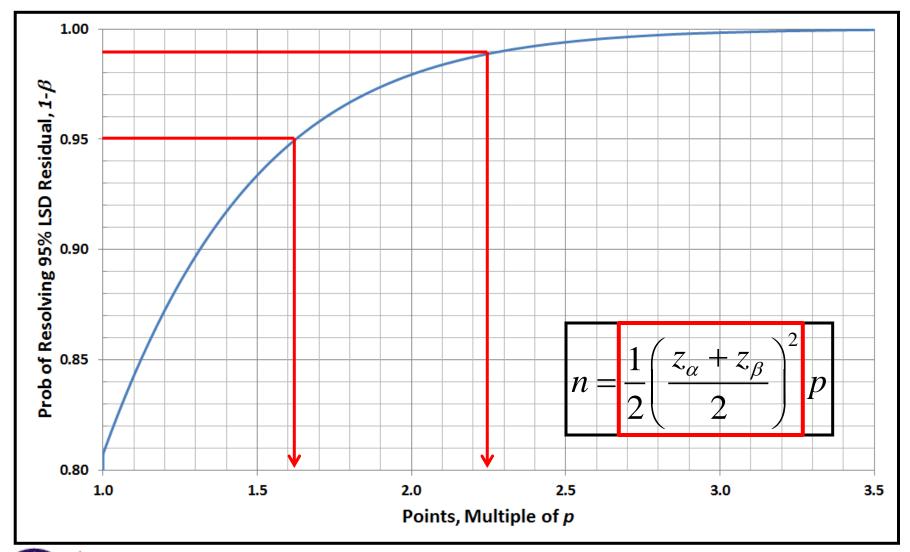
Model Term-Count Multiplier

Minimum to Resolve 95% LSD with α = 0.05



Model Term-Count Multiplier

Minimum to Resolve 95% LSD with α = 0.05



Another Special Case for Tolerance, δ

- Let δ = 95% PIHW (Prediction Interval Half-Width)
 - This is the smallest difference between a physical measurement and a <u>model prediction</u> that can be resolved with 95% confidence
 - It is a convenient tolerance spec because most curve-fitting software packages compute this automatically

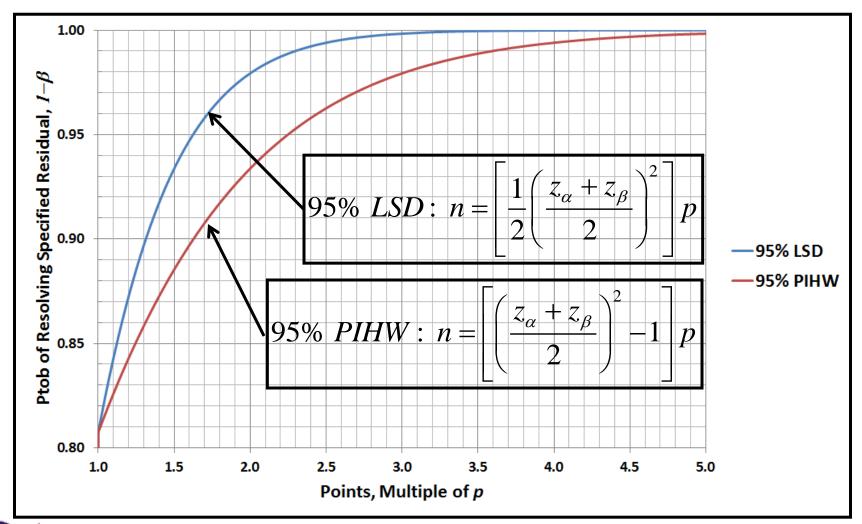
95% PIHW =
$$2\sqrt{1+\frac{p}{n}}\sigma \rightarrow K = 2\sqrt{1+\frac{p}{n}}$$

$$n = \left[\left(\frac{z_{\alpha} + z_{\beta}}{2} \right)^2 - 1 \right] p$$

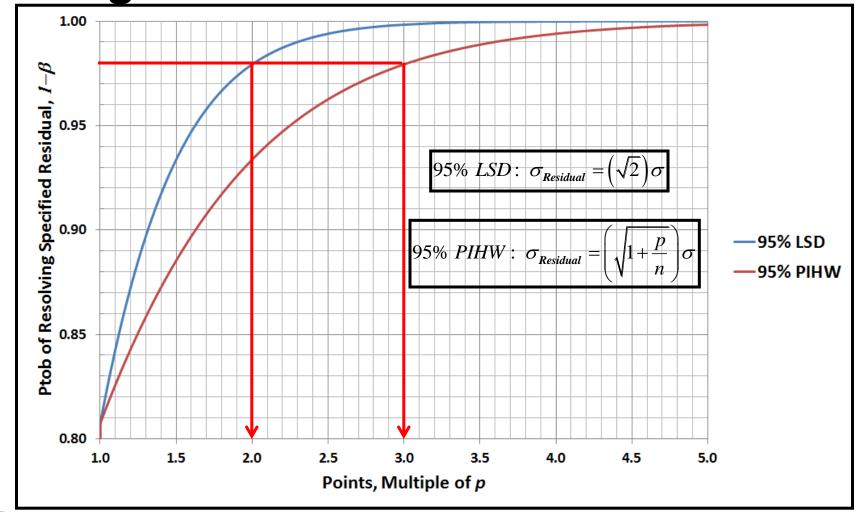


Model Term-Count Multiplier

Minimum to Resolve 95% LSD or 95% PIHW with α = 0.05



95% PIHW Tolerance Criterion is More Stringent than the 95% LSD Criterion



Numerical Scaling Example Typical OFAT Wind Tunnel Test

- Consider a wind tunnel test in which forces and moments are to be estimated as a function of four factors
 - Angles of Attack and Sideslip
 - Mach Number
 - Height (for ground effects)
- Typical OFAT levels might be as follows
 - AoA: -5° to +15° in 1° increments (21 levels)
 - Sideslip: 0° to +10° in 2° increments (6 levels)
 - Mach Number from 0.70 to 0.90 in 0.25 increments (9 levels)
 - Height (5 levels)
- Total of 21 x 6 x 9 x 5 = 5670 points (not atypical for OFAT test)
- Standard error in response estimate: σ



Numerical Scaling Example *Corresponding MDOE Scaling Case*

- Assume adequate fits can be achieved over three AoA subranges and two sideslip sub-ranges with 4th-order models
- A dth-order model in k factors has p terms (including intercept), where

$$p = \frac{(d+k)!}{d!k!} = \frac{(4+4)!}{4!4!} = 70$$

Assume a 95% LSD tolerance specification:

$$n = \frac{1}{2} \left(\frac{z_{\alpha} + z_{\beta}}{2} \right)^2 p$$

Numerical Scaling Example, Cont.

- Specify inference error risk tolerances
 - Max acceptable probability of rejecting a valid model: α = 0.05
 - Max acceptable probability of validating a bad model: β = 0.01
- Look up corresponding standard normal deviates, $z_{\alpha,\beta}$
 - For α = 0.05, z_{α} = 1.960 (double-sided null hypothesis)
 - For β = 0.01, $z_β$ = 2.326 (single-sided alternative hypothesis)
- Estimate data volume per subspace:

$$n = \frac{1}{2} \left(\frac{z_{\alpha} + z_{\beta}}{2} \right)^{2} p = \frac{1}{2} \left(\frac{1.960 + 2.326}{2} \right)^{2} 70 = 2.296 p = 161$$

Estimate total data volume (six subspaces): 6 x 161 = 966



MDOE/OFAT Comparison

There is a large apparent difference in OFAT and MDOE resource requirements

OFAT: 5670 points

- MDOE: 966 points

- The savings are not that dramatic, however
- MDOE methods invoke certain quality assurance tactics to defend against covariate effects
- Covariates are slowly varying, persisting factors that are not controlled by the experimenter
 - They are generally larger than ordinary random variations
 - They are not reproducible from test to test
 - They are largely overlooked in conventional OFAT testing



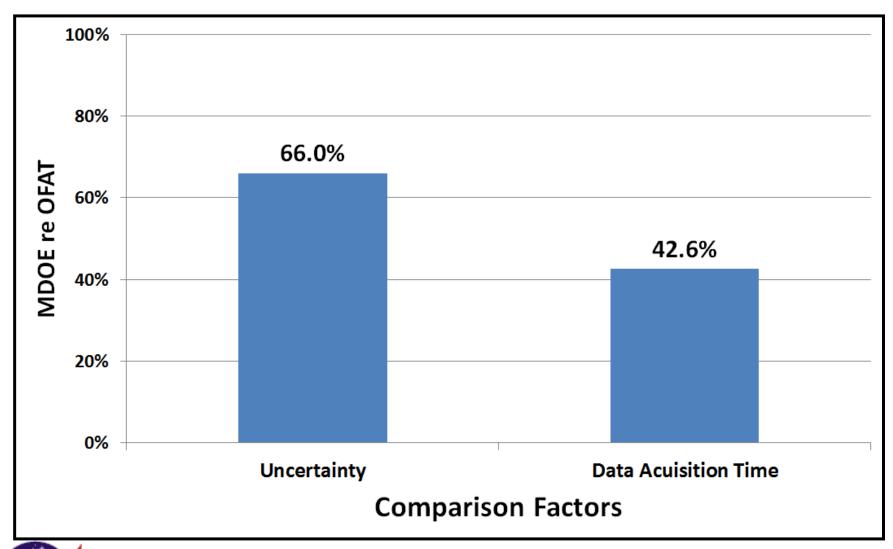
MDOE/OFAT Comparison, Cont.

- MDOE quality assurance tactics to defend against covariates cost a factor of 1.5 to 2.5 in data rate
 - In the time needed to acquire 966 MDOE points, up to 2.5 times as many OFAT points might be acquired
 - This would be 2.5 x 966 = 2415 points
- The MDOE data acquisition time is thus expected to be no more than a factor of 2415/5670 of the OFAT requirement, or 42.6% (and could be rather less)
- Note that the scaling resulted in a data volume requirement of n = 2.296p
- The MDOE standard error is thus

$$\sigma_{\hat{y}} = \left(\sqrt{\frac{p}{n}}\right)\sigma = \left(\sqrt{\frac{p}{2.296p}}\right)\sigma = 0.660\sigma$$



Quality and Productivity Comparison



Concluding Remarks

- Quality in wind tunnel testing is more properly expressed in terms of inference error probability than unexplained variance in the raw data
 - It is more important to get the right answer, than "good data"
 - This imposes a responsibility to articulate tolerance requirements
- There is a mathematical relationship between resource requirements and quality requirements
- Each new data point reduces inference error risk
 - Too little data means unacceptable inference error risk
 - Too much data means wasted resources
- The experimental aeronautics community might consider adopting the 95% LSD as a tolerance specification
- Then data volume in the range of 2 to 3 times the number of points needed to fit a model would typically be sufficient

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